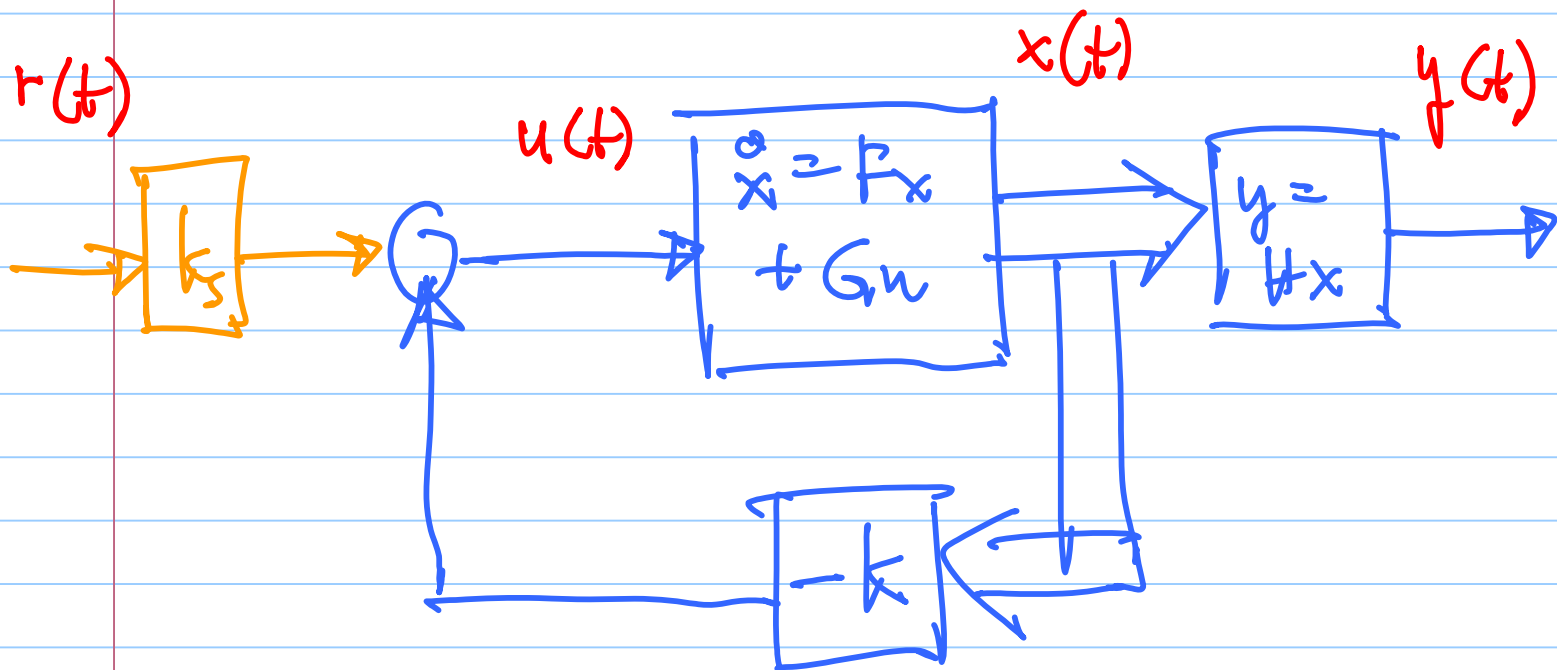
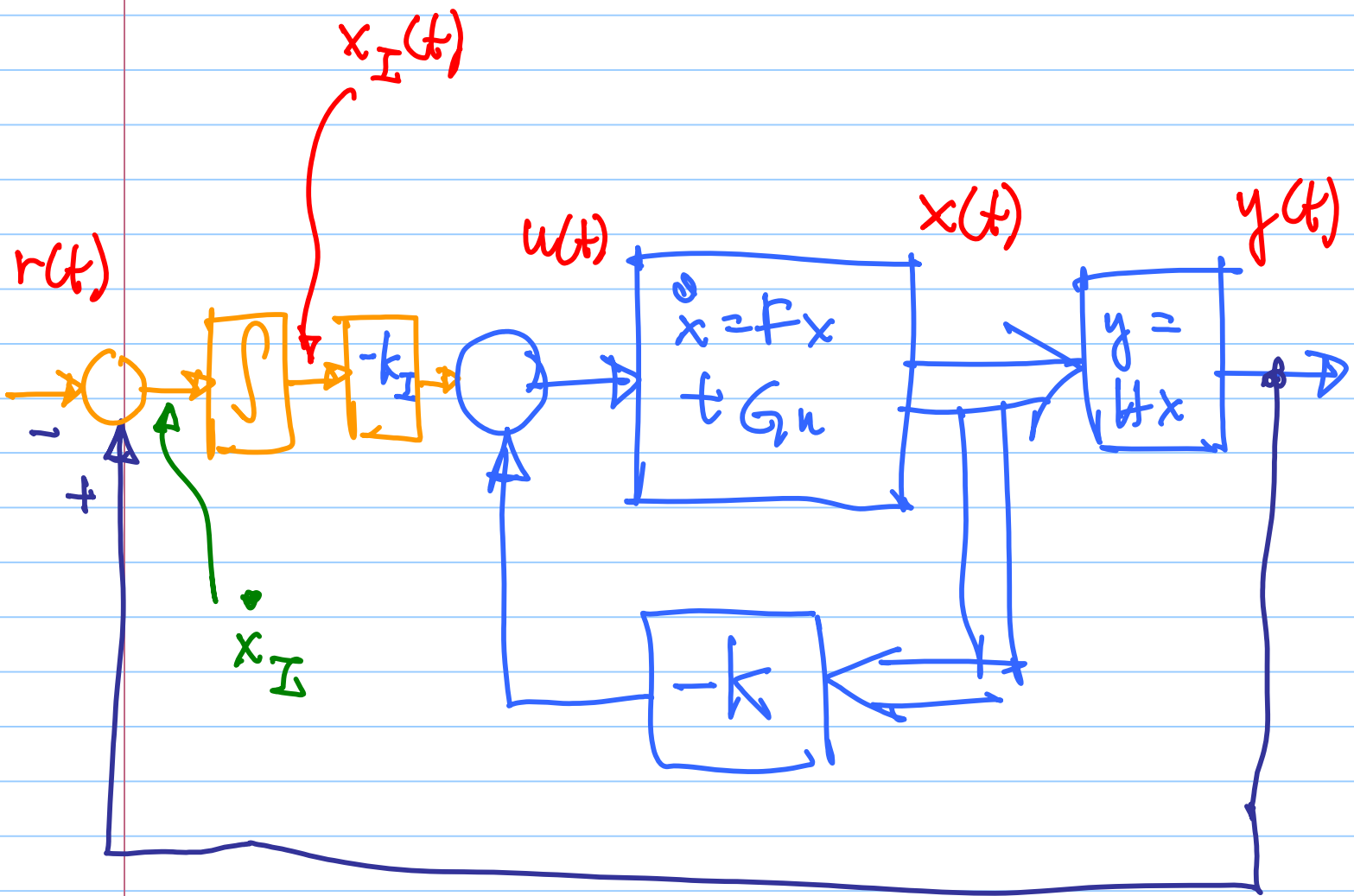


Our capability up to
this point is

6a
[Pre-
Class]





Next :

What happens if $x(t)$ is not fully measurable?

Going to the Lecture Notes

The approach is to estimate/observe the unmeasurable

$x(t)$ with the

"estimated/observed" signal
 $\hat{x}(t)$

Simple approach is to consider

firstly = "Open-Loop
Estimator/Observer"

... In the notes, shown clearly
that this simple approach
does not work well.

A much more proper approach
is to use the

"Closed-Loop Estimator/Observer"

$$\dot{x} = Fx + Gu$$

$$y = Hx \quad \text{--- (5.1)}$$

Not all the state-variables can be measured

So, we set up the "Closed-Loop Estimator/Observer"

given by:

$$\dot{\hat{x}} = F\hat{x} + Gu + L(y - H\hat{x})$$

How does this work?

Thus, consider the
"estimation error"

$$\tilde{x}(t) \triangleq x(t) - \hat{x}(t)$$

for this, we have:

$$\dot{\tilde{x}} = \dot{x} - \dot{\hat{x}}$$

$$= \{F_x + G_u\}$$

$$- \{F\hat{x} + G_u + L(y - H\hat{x})\}$$

Hx



$$\dot{\tilde{x}} = \{F - \underbrace{LH}_{\text{LH}}$$

(5.5)

Here, we need to choose L
so that the eigenvalues of
 $\{F - LH\}$
are sufficiently fast and
stable.

This is equivalent to requiring
the eigenvalues of

$$\{F - LH\}^T = \{F^T - H^T L^T\}$$

to be sufficiently fast and
stable.

Compare with

$$\{F - GK\}$$

where we chose an $\alpha_c(s)$!! matrix

($n \times n$)

Typical "Real" System

$$\dot{x} = Fx + Gu$$

$$y = Hx$$

Example could be the
d.c. motor in CA3

I Run some tests/calibrations
to obtain F , G and H

II If all state variables are
measurable, then we

can construct a controller

$$u = -Kx + k_s r$$

also
can use

$$\dot{x}_I = y - r$$

$$u = -Kx - k_I x_I$$

$$= -[K \ k_I] \begin{bmatrix} x \\ x_I \end{bmatrix}$$

(a) Choose desired

$$\alpha_c(s) = s^n + \alpha_1 s^{n-1} + \dots + \alpha_n$$

using

• Prototype Response
Tables

• SKL

• LQR

(b)

Calculate K using Ackermann's formula.

III

If state-variables are not all measurable, then have to construct an Estimator/Observer first.

→ i.e. construct

$$\hat{x} = F\hat{x} + Gu + L(y - H\hat{x})$$

Here, $\{F, G, H\}$ already obtained by calibration experiments. And Equation above will be implemented in real-time, e.g. using LabView software.

(a) For observer/estimator gain L , first choose desired

$$\overline{\alpha}_e(s) = s^n + \overline{\alpha}_1 s^{n-1} + \dots + \overline{\alpha}_n$$

using

- Prototype Response Tables
- SRL
- LQR

typically remembering to choose

bandwidth of $\bar{x}_e(s)$ to be about 3 to 5 times faster than $x_e(s)$.

(2) Calculate L using Ackermann's formula.

∴ we have

$$y = Hx$$

$$\dot{\hat{x}} = F\hat{x} + Gu + L(y - H\hat{x})$$

$$u = -K\hat{x} + k_s r$$

Summary

$$\dot{x} = Fx + Gu$$

$$y = Hx$$

I State - feedback

$$u = -Kx$$

leading to closed-loop

$$\dot{x} = (F - GK)x$$

$$y = Hx$$

Closed-loop poles are freely assignable
iff

$\mathcal{C}(F, G)$ is of full rank.

Recall:

$$\mathcal{C}(F, G) \triangleq [G; FG; F^2G; \dots; F^{n-1}G]$$

II Estimator

$$\dot{\hat{x}} = F \hat{x} + Gu + L(y - H\hat{x})$$

leading to estimator closed-loop

$$\tilde{x} \triangleq \hat{x} - x$$

$$\dot{\tilde{x}} = (F - LH)\tilde{x}$$

Estimator closed-loop poles are freely assignable iff

$\mathcal{O}(H, F)$ is of full rank.

Recall =

$$\mathcal{O}(H, F) \triangleq \begin{bmatrix} H \\ \vdots \\ H F \\ \vdots \\ H F^2 \\ \vdots \\ \vdots \\ \vdots \\ H F^{n-1} \end{bmatrix}$$

Summary

$$\begin{aligned}\dot{x} &= Fx + Gu \\ y &= Hx\end{aligned}$$

$$u = -Kx + k_s r$$

if all s-v.
are measurable.

Complete estimator-controller:

$$\dot{\hat{x}} = (F - GK - LH)\hat{x} + Ly$$

$$u = -K\hat{x} + k_s r$$

- State-fb gain K chosen so that

$$\alpha_c(s) = \det [sI - (F - GK)]$$

are the closed-loop state-fb poles.

- Estimator gain L chosen so that

$$\alpha_e(s) = \det [sI - (F - LH)]$$

are the closed-loop estimator poles.

→ Reduced-Order Observer...

Combined State-Feedback and State Estimator/Observer

$$\begin{aligned}\dot{x} &= Fx + Gu \\ y &= Hx\end{aligned} \quad \leftarrow (7.1)$$

Not all s.v. are measurable.

Thus, use:

$$\dot{\hat{x}} = F\hat{x} + Gu + L(y - H\hat{x}) \quad \leftarrow (7.2a)$$

$$u = -K\hat{x} \quad \leftarrow (7.2b)$$

Note that in above:

① K is chosen s.t. that

$$\det \{ sI - [F - GK] \} = \alpha_c(s) \quad \leftarrow (7.5a)$$

(II) L is chosen so that

$$\det \{ sI - [F - LH] \} = \alpha_0(s) \quad \text{--- (7.5b)}$$

In the above, the complete system is described by:

$$\begin{aligned} \dot{x} &= Fx + Gu \\ &= Fx + G \{ -K\hat{x} \} \end{aligned} \quad \text{--- (7.6a)}$$

$$\begin{aligned} \dot{\hat{x}} &= F\hat{x} + Gu + L(y - H\hat{x}) \\ &= F\hat{x} - GK\hat{x} + L(Hx - H\hat{x}) \end{aligned} \quad \text{--- (7.6b)}$$

and this is a complete description of the system with the full

equivalent state-variable
description from (7.6a)
and (7.6b) with:

$$\begin{bmatrix} \dot{x} \\ \hat{\dot{x}} \end{bmatrix} = \begin{bmatrix} F & -GK \\ LH & F - GK - LH \end{bmatrix} \begin{bmatrix} x \\ \hat{x} \end{bmatrix} \quad (7-7)$$

Thus, the above is one possible
complete description of the
system.

Consider next, the use of

$$\tilde{x} = x - \hat{x}$$

$$\text{i.e., } \hat{x} = x - \tilde{x}$$

then (7.6a) becomes:

$$\dot{x} = Fx - Gk(x - \hat{x}) \quad \text{--- (7.8a)}$$

and (7.6b) becomes:

$$\begin{aligned} \dot{\hat{x}} &= \{F - Gk\} \hat{x} + LH \{x - \hat{x}\} \\ &= \{F - Gk\} (x - \hat{x}) + LH \hat{x} \end{aligned} \quad \text{--- (7.8b)}$$

And also note from earlier:

$$\dot{\hat{x}} = \{F - LH\} \hat{x} \quad \text{--- (7.8c)}$$

Expand (7.8d) to:

$$\dot{x} = (F - GK)x - GK\hat{x} \quad \rightarrow (7.9a)$$

$$\dot{\hat{x}} = (F - LH)\hat{x} \quad \rightarrow (7.9b)$$

The equivalent complete overall system is thus:

$$\begin{bmatrix} \dot{x} \\ \dot{\hat{x}} \end{bmatrix} = \begin{bmatrix} F - GK & -GK \\ 0_{n \times n} & F - LH \end{bmatrix} \begin{bmatrix} x \\ \hat{x} \end{bmatrix}$$

\rightarrow F_{complete}

The eigenvalues of this complete $(2n \times 2n)$ system is given by =

$$\det \left\{ sI_{2n \times 2n} - F_{\text{complete}} \right\} = 0$$

$$\det \left[\begin{array}{c|c} sI_{n \times n} - \{F - G_k\} & G_k \\ \hline 0_{n \times n} & sI_{n \times n} - \{F - L_k H\} \end{array} \right] = 0$$

This simplifies to =

$$\det \left\{ sI_{n \times n} - [F - G_k] \right\} \cdot \det \left\{ sI_{n \times n} - [F - L_k H] \right\} = 0$$

$$i.e. \quad \alpha_c(s) \cdot \alpha_o(s) = 0$$

△△△

This is the so-called famous
"Kalman" Separation
Principle!!

$$\begin{bmatrix} x \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 & & \\ & - & \\ & - & \\ & & - \end{bmatrix} \begin{bmatrix} x \\ \hat{x} \end{bmatrix}$$

↙ T transformation